

One sample t-test

Hypotheses: $H_0: \mu = \#$

$$H_a: \mu \geq \#$$

test stat:

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

p-value:

$$P(t \geq \text{test stat} \mid df = n-1) =$$

confidence interval:

$$\bar{x} \pm t^* \cdot s/\sqrt{n} = (a, b)$$

from
table

Assumptions:

1) SRS

2) norm. pop.
or
 $n \geq 30$

Calculator:

T-test

T-Interval

Check

Matched Pairs t-test (one sample test)

* When? When we have 2 DEPENDENT samples that we want to combine (by subtracting) into 1 sample. *After-Before*

Hypotheses:

$$H_0: \mu_d = \# \quad (\text{often } 0)$$

test stat: $H_a: \mu_d \geq \#$

$$t = \frac{\bar{X}_d - \mu_d}{S_d / \sqrt{n_d}}$$

p-value:

Same

Assumptions:

- 1) SRS
- 2) norm. pop. of differences
or
 $n_d \geq 30$

Calculator:

T-test
T-Int.

confidence interval:

~~write~~ Interpret:

$$\bar{X}_d \pm t^* \cdot S_d / \sqrt{n_d}$$

... that the mean diff. of ____ is
btw. a & b

Two sample t-test

Hypotheses: $H_0: \mu_1 = \mu_2$

$$H_a: \mu_1 \neq \mu_2$$

Assumptions:

1) 2 indep. SRS

2) 2 normal pop.
or

$$n_1 \text{ and } n_2 \geq 30$$

Calculator:

test stat:

$$t = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}}$$

p-value:

$$P(t \geq \underline{\hspace{2cm}} \mid df = \text{on calc.}) = tcdf(LB, UB, df)$$

confidence interval & interpretation: ^{or smaller of $n_1 - 1$ or $n_2 - 1$}

$$(\bar{X}_1 - \bar{X}_2) \pm t^* \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}} = (a, b)$$

We are % conf that the diff. b/w. the means of #1 & #2
is b/w a and b. *O

Two sample t-test pooled ☐

when? when $\sigma_1 = \sigma_2$ (pop. std. dev.)

What is sp? pooled std. dev.

on calc. pooled estimator of σ

test stat:

$$t = \frac{\bar{X}_1 - \bar{X}_2}{S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

confidence interval:

$$(\bar{X}_1 - \bar{X}_2) \pm t^* S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

$$df = n_1 + n_2 - 2 \text{ (on calc.)}$$

Assumptions:

check

1)
2)

3) $\sigma_1 = \sigma_2$

1) stated/
2) $S_1 \approx S_2$

Calculator:

Pooled- YES

t

* unknown σ
estimate w/ s

• df

• t-distribution

$N(0, ?)$

depends on
df/n

• t wider than z

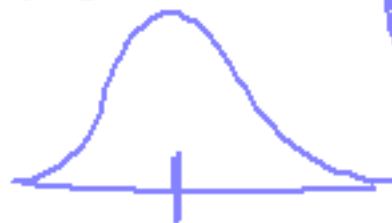
• as $n \uparrow$, t become closer to z

Sim

• t with $df = \infty$,

$$t = z$$

• normal shape



• testing means

z

• known σ

• Std. Normal Distr.
 $N(0, 1)$

